







# Decision Tree Algorithm







## **EXAMPLE**

- What are some good predictors of whether you will stay in or go out for this weekend?
- Factors: Weather, your spouse in town, important game on TV, do you have something urgent to do at work.
- All these factors usually have some order of importance.







#### THE WEATHER IS IT RAINING? THIS IS EXACTLY WHAT A YES DECISION TREE STAY IN DO YOU HAVE SOMETHING URGENT TO DO AT WORK? LOOKS LIKE NO YES STAY IN IS THERE AN IMPORTANT GAME ON TV? YES IS YOUR SPOUSE/SIGNIFICANT STAY IN OTHER IN TOWN? NO YES STAY IN GO OUT







# WHAT ARE DECISION TREES?

- In Machine Learning decision tree predicts the outcome given the values of the input variables.
- It's a Supervised learning algorithm.
- Decision Tree helps in making decision in structured manner and visually represents all the variables involved and the consequences in each case.







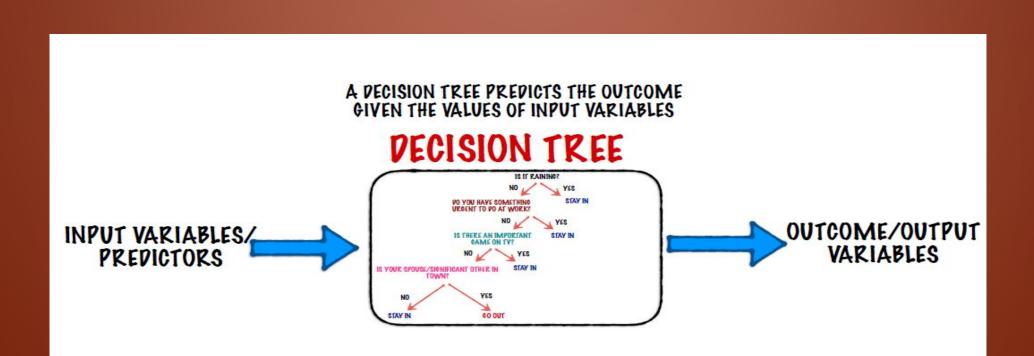
## Advantages and Disadvantages

- 1. Easy to interpret and perfect for visual representation.
- 2. It works both with categorical and numerical features.
- 3. Nonlinear parameters don't affect its performance.
- 4. Little effort is required in data preparations.

- 1. Overfitting as algorithm tends to pick up the noise in the data.
- 2. High Variance as model tends to get unstable with little variation in data.
- Low bias which makes the model difficult to work with new data.













## **DECISION TREE**





OUTCOME/OUTPUT VARIABLES

# DECISION TREE LEARNING

RECURSIVE PARTITIONING

IS THE MOST COMMON STRATEGY FOR DECISION TREE LEARNING

IS THE PROCESS OF CREATING/LEARNING A DECISION TREE FROM TRAINING DATA

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ALGORITHMS BASED ON RECURSIVE PARTITIONING







## **Keywords of Decision Tree**

- The topmost feature would be called as root node
- Then we have other features as branches
- The endpoints are called **leaf nodes**.









Entropy: The measure of randomness and unpredictability in the dataset.

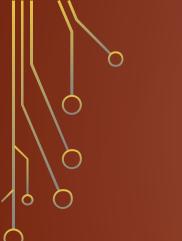
$$E = -\sum_{i}^{C} p_i \log_2 p_i$$

Information Gain is measure of entropy after the dataset has been split.

Range(0 to 1)
$$Gain(S, A) = Entropy(S) - \sum_{v \in Values(A)} \frac{|S_v|}{|S|} Entropy(S_v)$$

Gini Impurity is used for calculation of purity of a split.

$$G = \sum_{i=1}^C p(i)*(1-p(i))$$









## How the Decision Tree works

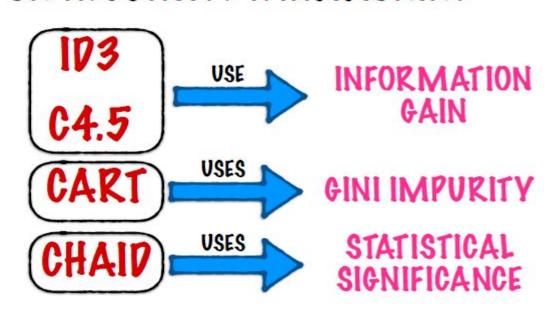






# DECISION TREE LEARNING ALGORITHMS BASED ON RECURSIVE PARTITIONING

EACH HAS A SLIGHTLY DIFFERENT WAY OF ARRIVING AT THE BEST ATTRIBUTE (OR) MEASURING THE HOMOGENEITY OF A SUBSET









# FRODO AND SAM ATE AT A RESTAURANT EVERY DAY LAST WEEK AND RATED IT ON EACH DAY

MONDAY	GOOD	
TUESDAY	BAD	
WEDNESDAY	GOOD	
THURSDAY	GOOD	
FRIDAY	GOOD	
SATURDAY	BAD	
SUNDAY	GOOD	

## AT THE END OF THE WEEK,

#### FRODO SAYS

THE FOOD IS GOOD AT THIS RESTAURANT

#### SAM SAYS

THE FOOD IS GOOD AT THIS RESTAURANT ON ALL DAYS EXCEPT TUESDAYS AND SATURDAYS

WHICH ONE OF THEM IS RIGHT?





FRODO'S



SAM'S

## WHICH ONE OF THEM IS RIGHT?

	TRAINING SET	MODEL	MODEL
MONDAY	GOOD	GOOD	GOOD
TUESDAY	BAD	GOOD	BAD
WEDNESDAY	GOOD	GOOD	GOOD
THURSDAY	GOOD	GOOD	GOOD
FRIDAY	GOOD	GOOD	GOOD
SATURDAY	BAD	GOOD	BAD
SUNDAY	GOOD	GOOD	GOOD

WE COULD CHECK EACH OF THEIR STATEMENTS AGAINST THE DATA WE ALREADY HAVE 71% 100%

**ACCURACY** 







## WHICH ONE OF THEM IS RIGHT?

FRODO'S MODEL SAM'S MODEL

MONDAY	GOOD	GOOD	GOOD
TUESDAY	BAD	GOOD	BAD
WEDNESDAY	GOOD	710	1000
THURSDAY	GOOD	71%	100%
FRIDAY	GOOD	GOOD	GOOD
SATURDAY	BAD	GOOD	BAD
SUNDAY	GOOD	GOOD	GOOD
		71%	100%

ON THE TRAINING SET, FRODO'S MODEL HAS 71% **ACCURACY AND SAM'S** MODEL HAS 100% ACCURACY

FROM THIS, IT SEEMS LIKE SAM'S MODEL IS BETTER.

ACCURACY

SAM AND FRODO GO BACK TO THE RESTAURANT NEXT WEEK







## WHICH ONE OF THEM IS RIGHT?

FRODO'S MODEL SAM'S MODEL

MONDAY	GOOD	GOOD	GOOD
TUESDAY	BAD	GOOD	BAD
WEDNESDAY	GOOD	GOOD	GOOD
THURSDAY	GOOD	171%1	100%
FRIDAY	GOOD	GOOD	GOOD
SATURDAY	BAD	GOOD	BAD
SUNDAY	GOOD	GOOD	GOOD
MONDAY	GOOD	GOOD	GOOD
TUESDAY	GOOD	GOOD	BAD
WEDNESDAY	BAD		1000
THURSDAY	GOOD	71%	42%
FRIDAY	GOOD	GOOD	GOOD
SATURDAY	GOOD	GOOD	BAD
SUNDAY	BAD	GOOD	GOOD

ON THE TRAINING SET,
FRODO'S MODEL HAS 71%
ACCURACY AND SAM'S
MODEL HAS 100%
ACCURACY

SAM AND FRODO GO BACK TO THE RESTAURANT NEXT WEEK

ON NEW DATA, FRODO'S
MODEL HAS 71%
ACCURACY AND SAM'S
MODEL HAS 42%
ACCURACY

WEEK 2

WEEK 1



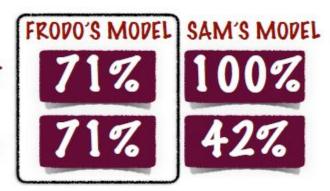




## WHICH ONE OF THEM IS RIGHT?

TRAINING SET

**NEW/UNSEEN DATA** 



WHAT HAPPENED HERE?

FRODO'S MODEL IS THE BETTER MODEL

IT GENERALIZES WELL

FRODO'S MODEL
PERFORMS WELL ON
BOTH TRAINING AND
NEW/UNSEEN DATA







## WHAT HAPPENED HERE?

FRODO'S MODEL

SAM'S MODEL

TRAINING SET

71%

100%

NEW/UNSEEN DATA

71%

42%

THE FOOD IS
GOOD AT THIS
RESTAURANT
ON ALL DAYS

EXCEPT TUESDAYS AND SATURDAYS FRODO'S MODEL IS SIMPLER ("DUMBER", IN FACT), YET IT PERFORMS BETTER

SAM'S MODEL IS MORE COMPLEX, AND MORE ACCURATE ON THE TRAINING SET

YET, IT PERFORMS BADLY ON NEW DATA IE, SAM'S MODEL DOES NOT GENERALIZE WELL







# THE FOOD IS GOOD AT THIS RESTAURANT ON ALL DAYS EXCEPT TUESDAYS AND SATURDAYS

SAM'S MODEL PICKS UP ON A RELATIONSHIP BETWEEN THE WEEKDAY AND THE QUALITY OF FOOD

THIS RELATIONSHIP
HOWEVER, IS SPECIFIC TO THE
TRAINING SET, AND NOT TRUE
IN GENERAL

SAM'S MODEL IS A PERFECT EXAMPLE OF

# OVERFITTING

OVERFITTING OCCURS WHEN A MODEL PICKS UP ON RANDOM PHENOMENA OR NOISE PRESENT IN THE TRAINING SET INSTEAD OF THE UNDERLYING RELATIONSHIP BETWEEN THE INPUT AND OUTPUT



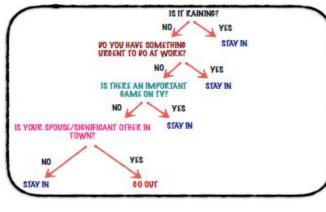




A DECISION TREE CAN BE USED TO SOLVE MACHINE LEARNING PROBLEMS

A DECISION TREE
PREDICTS THE
OUTCOME GIVEN THE
VALUES OF INPUT
VARIABLES

INPUT VARIABLES/
PREDICTORS



OUTCOME/OUTPUT VARIABLES

DECISION TREES ARE VERY PRONE TO THE RISK OF OVERFITTING

ENSEMBLE LEARNING CAN MITIGATE THE RISK OF OVERFITTING









## **RANDOM FOREST ALGORITHM**

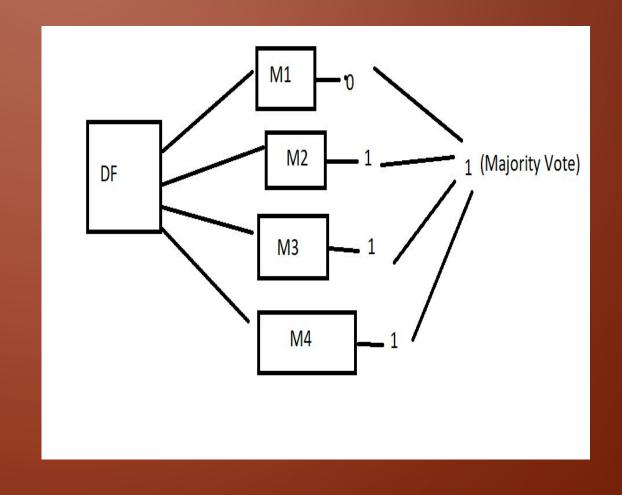






# Bagging (Ensemble Technique)

- Also known as Bootstrap
   Aggregation.
- Row Sampling with
   Replacement is performed
   (Bootstrap).
- Each Model trains and predicts an outcome and the majority is taken as final output (Aggregation).







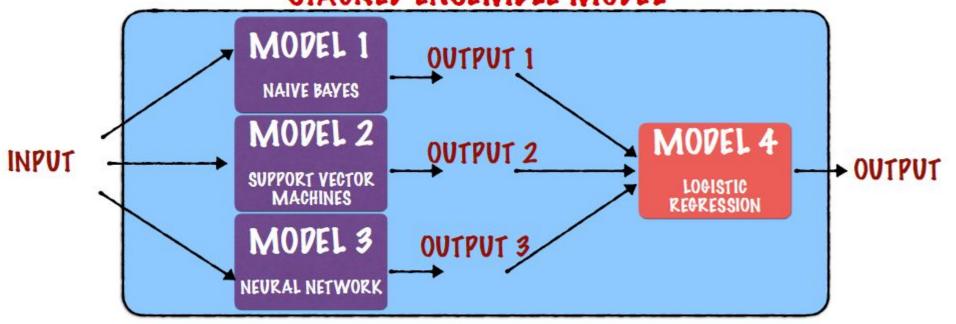


# STACKING

(AKA BLENDING AKA STACKED GENERALIZATION)

INVOLVES USING A MACHINE LEARNING APPROACH TO COMBINE THE RESULTS OF THE ENSEMBLE MEMBERS

#### STACKED ENSEMBLE MODEL









## **Random Forest**

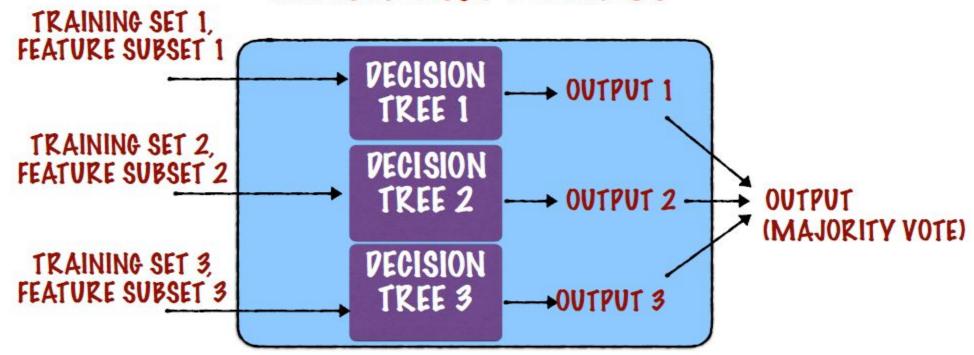
- Random forest, like it names implies consists of large number of individual decision trees that operate as an ensemble.
- Each individual tree has spits out a class prediction and class with most votes becomes our final model prediction.
- In Random Forest row sampling and feature sample is used which makes its a powerful algorithm.







# RANDOM FOREST









#### **Application of Random Forest**



Remote Sensing

Used in ETM devices to acquire images of the earth's surface.

Accuracy is higher and training time is less



**Object Detection** 

Multiclass object detection is done using Random Forest algorithms

Provides better detection in complicated environments



Kinect

Random Forest is used in a game console called Kinect

Tracks body movements and recreates it in the game







## **Why Random Forest?**



No overfitting

Use of multiple trees reduce the risk of overfitting

Training time is less



High accuracy

Runs efficiently on large database

For large data, it produces highly accurate predictions



Estimates missing data

Random Forest can maintain accuracy when a large proportion of data is missing







# **Boosting** (Ensemble technique)

- ADA Boost (Adaptative Boost)
- Gradient Boost
- XGB Boost

# Performance Metrics







## ROC and AUC Curve

ROC: Receiver Operating Characteristic is mostly used to visualize binary classifier.

AUC: Area Under Curve is measure of ability of a classifier to distinguish and be the summary of ROC curve.

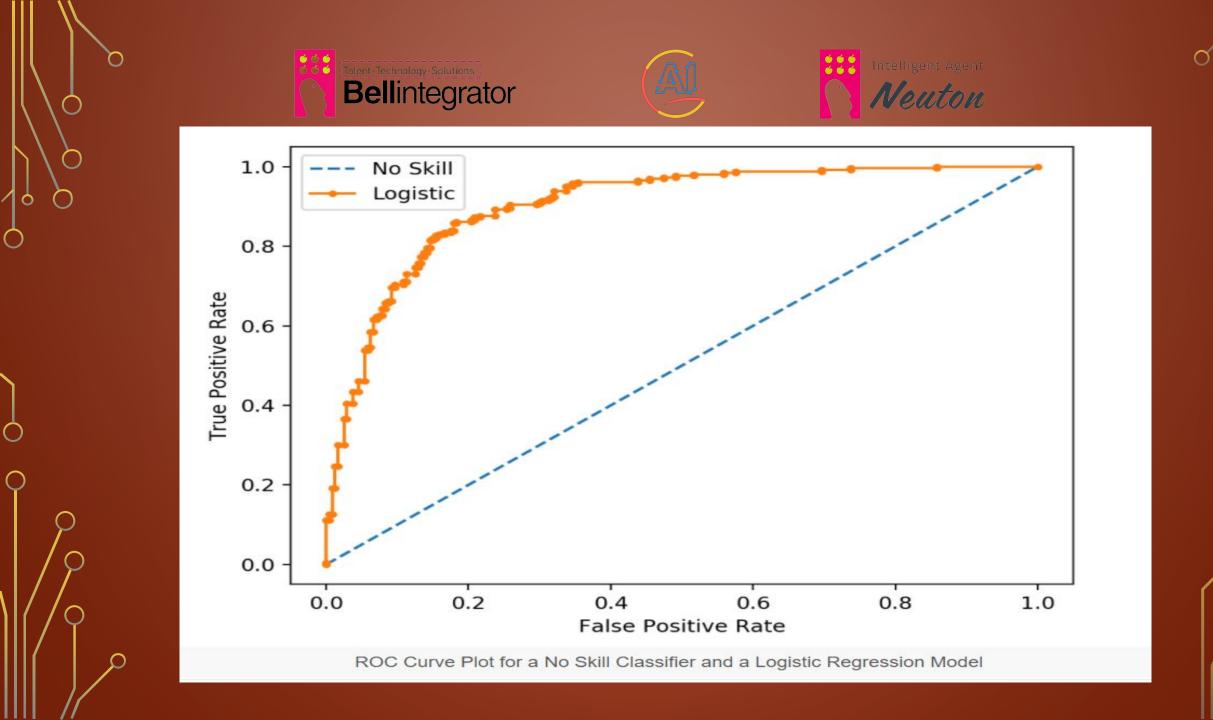
The ROC is plotted using the True Positive Rate and the False Positive Rate.







- Its plotted is given as Sensitivity vs 1- Specificity.
- Sensitivity is True Positive Rate which is ratio of True positives and sum of True positives and False negatives.
- Specificity is False Positive Rate which is of False Positives and sum of True negatives and False positives.









# Industry Use cases of Decision Trees

- Commonly used in data mining.
- Loan Approval
- In Finance sector, forecasting future outcomes and assigning probabilities to those outcomes







# Industry Use cases of Random Forest

- Banking Sector: Banking sectors consists of most users. Used for Fraud
   Detection
- Medicines needs complex combination of chemicals. Random forest comnes handy in this scenario.
- Stock Market: Stock behaviour analysis can be done using Random Forest.







## Summary

- Decision Tree model is like a white box.
- Decision tree performs well in case of non linear data.
- Decision tree tends to overfit.
- Overfitting can be mitigated using various Ensemble techniques
- Random Forest is an ensemble of decision trees.
- Random forest widely popular because of its missing value handling and no overfitting.

